Modeling Intuition as a System

"To explain the integration of information, we need only exhibit mechanisms by which information is brought together and exploited by later processes."   
**-David Chalmers, Facing Up to The Problem of Consciousness**

# Introduction

## Objective

General intelligence of the type we possess exists exclusively, for now, in the domain of the conscious human. Therefore, an understanding of the mechanisms leading to our conscious experience may be required if an artificial general intelligence is to one day be realized.

One defining aspect of human intelligence is our ability to subconsciously form new connections between abstract concepts, which then seem to "bubble up" to the forefront of our attention. This phenomenon, commonly called intuition, is responsible not only for our most startling and profound "Aha!" moments, but also for the seemingly arbitrary changes in our awareness of, say, the ticking of a clock on the wall. And although intuition is, unfortunately, a system that exists inside us as a “black box” (we have no conscious access to its decision-making process), the ways that we experience these shifts of attention unwillingly and "out of the blue" provide powerful clues to its underlying mechanisms.

With that in mind, the purpose of this project is to explore these ideas by developing an ensemble learning system (which we will call an *agent*) who's “conscious awareness” (its *attention*) receives input according to some optimization function (an *intuition)* with the goal of learning (*evolving)* to filter abstract concepts from search-space (*environmental*) noise. In addition, noting the hierarchal nature of information, the agent was designed to scale from a single agent, to a node-agent in a network of such agents, thereby facilitating the bootstrapping of an increasingly advanced intuition.

## Technologies

The framework and agent were developed in the Python (3.6) programming language, chosen for its rapid prototyping capabilities and reflective nature. 3rd party library usage includes KarooGP (genetic programming) and PyTorch (machine learning).

# Problem Domain

To facilitate exploration of this topic, a problem domain was chosen based on the realization that the agent’s environment should, ideally:

1. Afford the agent an opportunity to explore a complex, unknown search-space.
2. Be multi-context.
3. Provide mechanisms for signaling feedback to the agent.
4. Have the potential for practical application.

With that in mind, the agent was applied to the task of learning the Python programming language and, eventually, accomplishing some goal using it. This meets our criteria very well, because Python -

1. Is a complex space in which an isolated learning environment may be constructed.
2. Provides various contexts. Ex: keywords, functions, arguments, classes, instances, programs, etc.
3. Exposes methods such as keyword.iskeyword(s) and callable(s), enabling the agent to “ask” if some string s represents a Python keyword or a callable function, respectively. Further, because the agent itself is written in Python, custom methods may be written for the agent to provide more specific feedback, such as is\_python(s) which returns True if s represents a valid Python program with no syntax errors that generates no exceptions.
4. Is highly reflexive, allowing a sufficiently advanced agent to, quite literally, write a version of itself that solves some arbitrary problem. Indeed, even writing its own feedback mechanisms for querying fitness heuristics.

# Modeling Intuition

## Design Considerations

Subjectively (from the perspective of our awareness) experiencing intuition “feels” no different than experiencing input from any of our garden-variety five senses except for (at least) one notable difference: intuitive input carries with it contextual meaning and symbolic comprehension about our environment - ideas composed by filtering environmental input through the sieve of one’s accumulated life experience. In this way it can be thought of as a sixth sensory input channel, different from the first five in that it serves information from our sub-conscious.

Our agent was conceived as a model representing this interpretation of intuition, and designed based on the following observations of intuitive behavior in humans -

* Observation 1  
  a) We are capable of reacting to events faster than we have time to logically determine a rational course of action(CITE). Regardless, we may still make very good “in-the-moment” approximations that seem to involve no conscious thought.  
    
  b) We can articulate the rules we use to solve a given problem but are generally unable to explain why we chose to consider one specific set of rules over another (Pitrat, 2010).  
    
  c) Shifts in changes to our awareness are often (if not always) autonomic (CITE). For example, we cannot dictate when a song will become stuck in our head, or which clock we can suddenly hear ticking.   
    
  \* Conclusion 1  
  Some system operating below our level of consciousness exists for selectively serving information into our awareness.
* Observation 2  
  Seemingly trivial and/or unproductive “mistakes” often come into our awareness. For example, songs DO get stuck in our head and we DO become suddenly aware of a ticking clock for no apparent reason. Contrapositively, we’re often oblivious to important environmental queues, especially when properly distracted (a trait commonly exploited by magicians and pick pockets alike).  
    
  \* Conclusion 2   
  Intuition is not perfect. However, "mistakes” have evolutionary value (e.g. genetic mutation compels biological adaptation). Therefore, as a biological system itself, it is reasonable to assume the mechanism by which it learns to optimally processes and disseminate information is Darwinian in nature.
* Observation 3  
  The state of our awareness affects our intuition. Contrarywise, the state of our intuition affects our awareness. To demonstrate this consider what occurs when focusing one’s awareness on a particular topic; we become acutely aware of *it* and much less aware of everything *else* – apparently dictating either the amount of work performed on, or the priority assigned to, specific information related to the subject of our intense interest.  
    
  \* Conclusion 3  
  Awareness and intuition exist as a feedback loop, each guiding each other in lockstep.
* Observation 4  
  It is human nature to thoroughly explore our environment for personal gain. As evidence of this, I offer humankind’s domination of its natural habitat and technological innovation.  
    
  \* Conclusion 4  
  An environmentally-aware agent, motivated by evolutionary forces and possessing an intuition, will naturally act to explore its environment and actively develop mechanisms for its exploitations.

## The Model

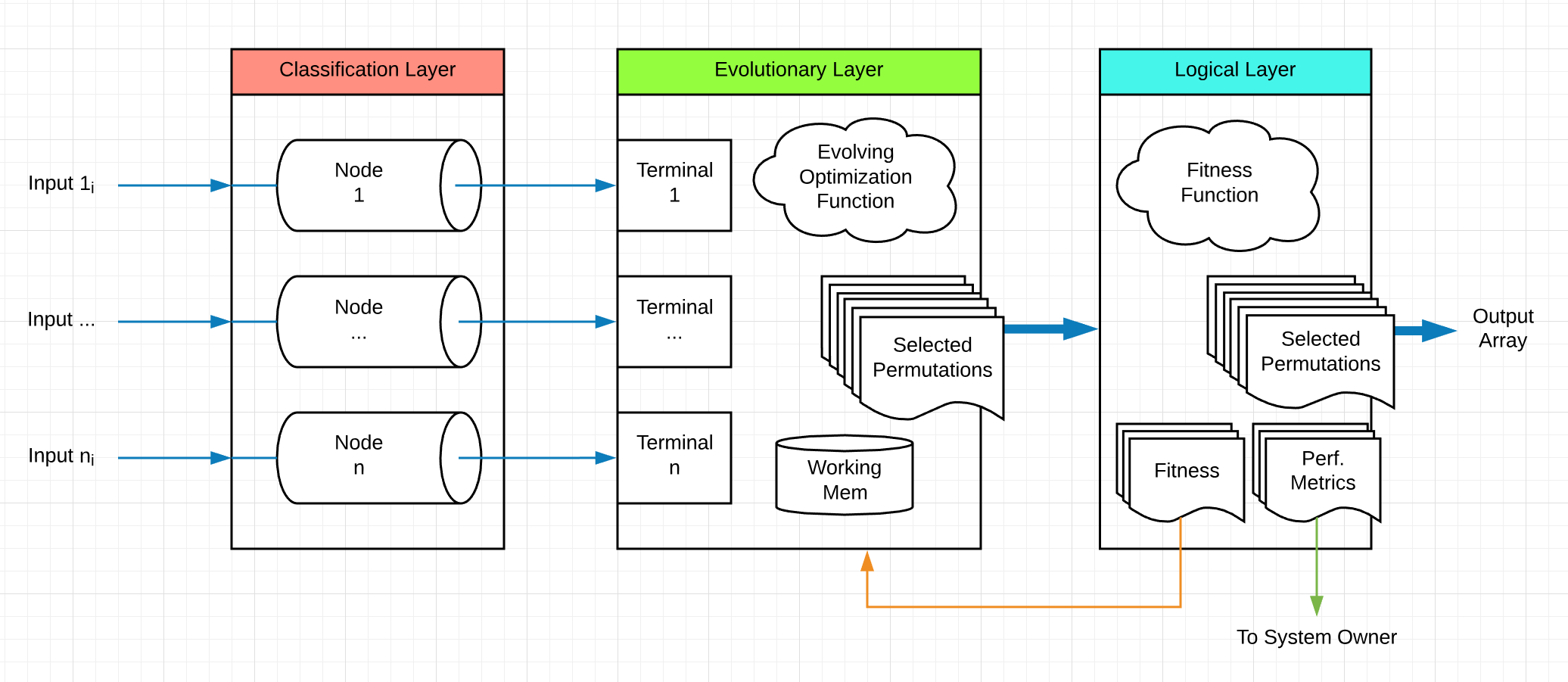


Figure 1

The intuitive model consists of 4 layers, labeled  *Classifer, Evolutionary,* and *Logical*. Data is mostly feed-forward, with channels existing for the back-propagation of layer-state and contextual fitness.

**Note:** Several models were initially conceived (see: Other Approaches/Models)

Operation proceeds, in general, in the following way

1. Classifying it.
2. and disseminating it in an optimal way to the **.** consists as described below, and given by Figure 1.

### Layer 1 – The Classifier Layer

The classifier layer is an **input layer** and representsour ability to rapidly (intuitively) classify sensory input based on previous exposure to similar patterns of input.

It is as an arrangement of parallel classifier **nodes** of **depth** d, where the input to each node is a sample of some subset of the agent's environment, and the output of each node is its classification of that input.

This layer might be thought of as an input bus, and each node a single line on it.

**Implementation**:   
Each classifier-layer node is implemented as an Artificial Neural Network (ANN) of the following shape -.

…

**Training / Validation:**Each node is trained independently (though still in parallel) in a supervised and offline fashion. Training/validation data is not constrained to share similarities across nodes. On the contrary, disjoint sets will encourage linear separability in subsequent layers.

**Output:**Each node’s classification is pushed to that node’s output channel for use as input to the subsequent layer.

### Layer 2 – The Evolutionary Layer

**The intuitive layer represents** …

On each state change, each equation evolves in an online manner according to the genetic algorithm, the fitness for which is provided as feedback from the subsequent layer

* How is the genetic algorithm diff than just naother ann? Does it prevent overtraining

In this way, the agent's "intuition" learns how to best allocate the agent's "attention" while allowing "mistakes" to enter its awareness. These mistakes represent possible new conceptual connections across classifier-layer nodes

This layer also handles representation of a short-term, finite, “working memory”, allowing us to hold previous environmental symbols in our minds. The agent’s “memory” represents this and allows introduction of error into the connection-forming process.

### Layer 3 – The Logical Layer

**The logical layer represents** .. .and examines the output-nodes of the intutive layer to draw concl

# The Agent

* The agent exists in the context of its environmental inputs and “intuitively” learns the symbols present in it. In this way, it can be thought of as an automatic tokenizer.
* Context - an amalgamation of input from the five-senses input - we listen and see at the same time?

## The data

Extracted w/Pandas

Why this set?

* Familiar with it.
* Deep learnable but w/Simplicity - No need for conv or pooling of layers (though agent is capable of doing so).

Data Set desc: <https://archive.ics.uci.edu/ml/datasets/Letter+Recognition>

## Fitness Heuristics

* What heuristics might drive it and its exploration - neophilia? Self-actualization? Guilt?

## Agent operation/training/parameters/algorithms/methodology

“Step” – A Walkthrough

Tunable Constants…

# Performance

* Human context switching is approx 200 ms. 20-50ms is reasonably real-time.

# Failed Approaches/Models

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# What’s Next

Bootstrapping:

* Each agent represents a context

Branching/Bootstrapping:

1. Branch into a hierarichal structure of agents (faciliated by reflection) – each representing one context. Contexts are not necessarily predefined.
2. L2 branch after some number of increases of the last x iters
3. Each agent represents a single "concept" such as a letter, or a python kwd
4. If a branch agent does not learn enough over some t, rm it (log to kb?) - it's inputs do not form any concepts
5. One node has one goal, a clique has another goal, a network has an even still more encompassing goal (ex: find free memory?)

L2 Tuning:

1. Monitor accuracy heuristics over time -
2. increase max pop size after some accuracy threshold
3. decrease max pop size if proprtion of unfit to fit outputs calls for it

More heuristics:

* We encounter many heuristics in life
* OBSERVATION: We are not always compelled to act rationally - in addition to logical reasoning, emotional (irrational) reasoning strongly influences our behavior.   
   **CONCLUSION 3** (Also in correlation w/CONCLUSION 2):  
  The heuristics guiding our intuition’s evolutionary journey may be associated with irrational drivers such as guilt, loneliness, boredom, etc.
  + Innate drive based on environmental queues/heuristics comples us toward the new, and
  + How are the inputs wighted? Biased by "amount" of input and log(type)? I.e. Fire vs. TV - Fire is big and new.
  + What drives the shifts?
  + The genetic alg decides attentive allocation between input data elements. Heuristics?

# Appendix A:

## Influences

Nicolai Tesla, Daniel Hofstadter, Ray Kurzweil, Richard Dawkins, Jacques Pitrat, Danial Kahneman

# References

Pitrat, J. (2010). Artificial beings Wiley-ISTE.

# Appendix B: Technical Information

## Usage Instructions

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CLI –

Tunable constants -